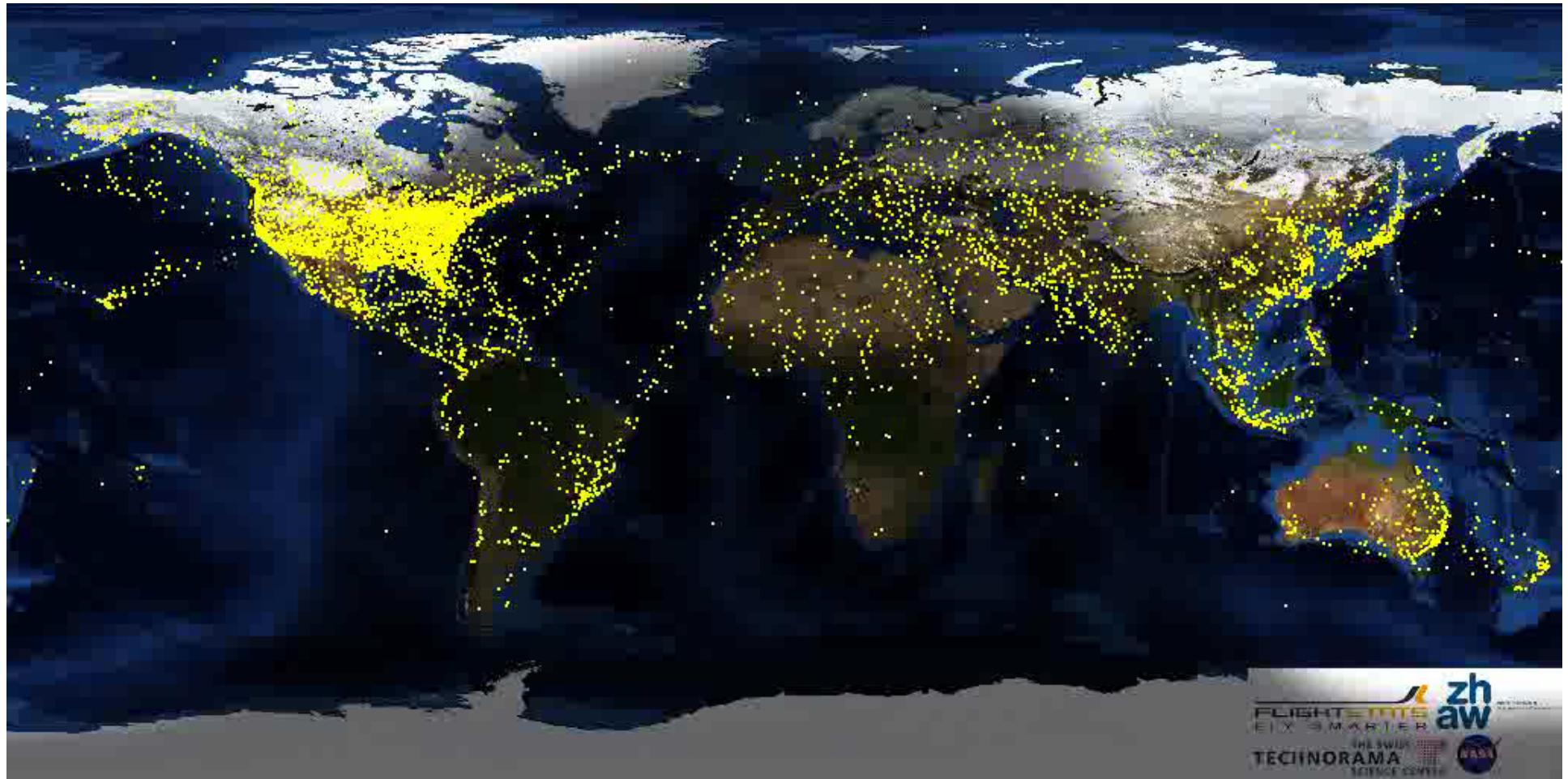




Towards Verification and Validation for Increased Autonomy

Dimitra Giannakopoulou

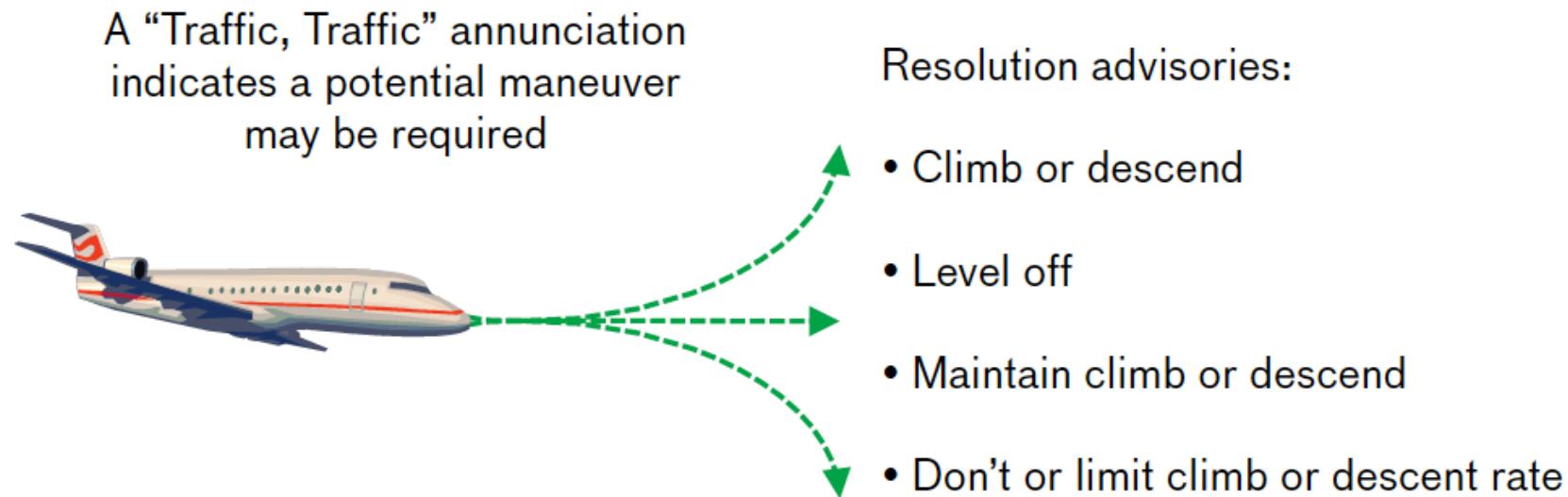
an aircraft is not alone in the sky...



FLIGHTS
EIX SMARTER
TECHNORAMA
zhaw
THE SWISS
SCIENCE CENTER
NASA

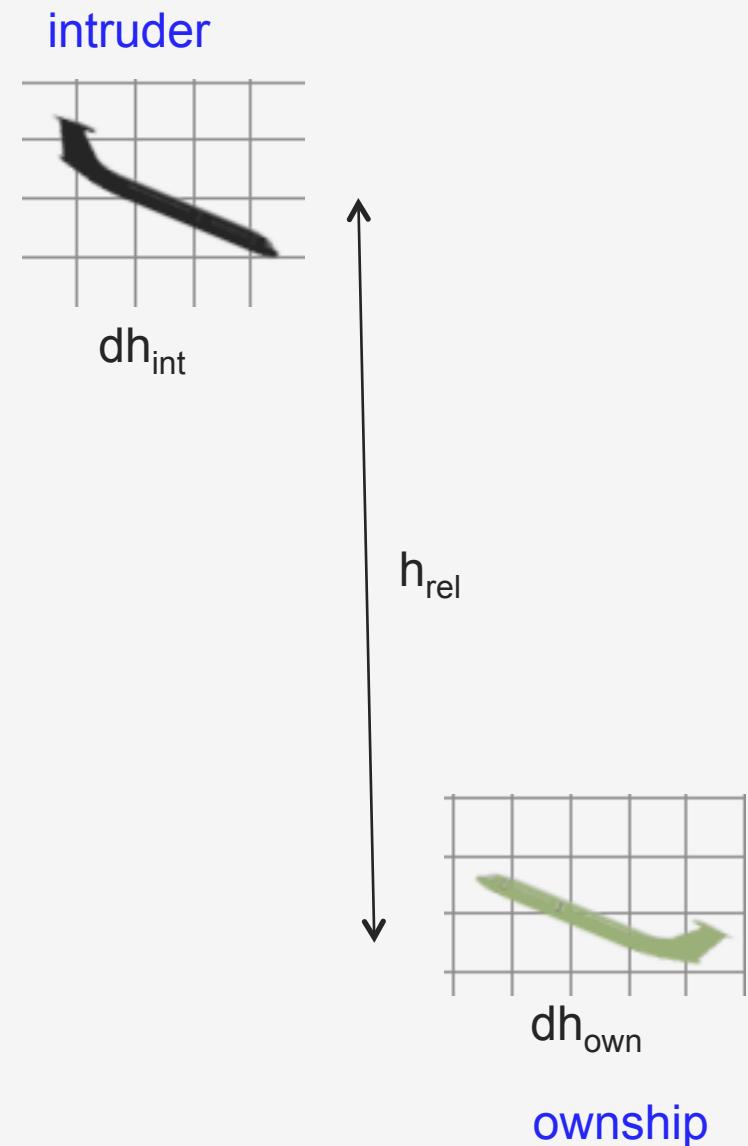
Safe Air Transportation

- Air traffic operations are expected to increase significantly.
Automation must maintain or exceed current safety standards
- Separation Assurance – algorithms and systems gradually taking the role of air-traffic controllers to enable reduced aircraft separation
- Onboard-Collision Avoidance Systems – TCAS, ACAS X



ACAS X – a completely new paradigm

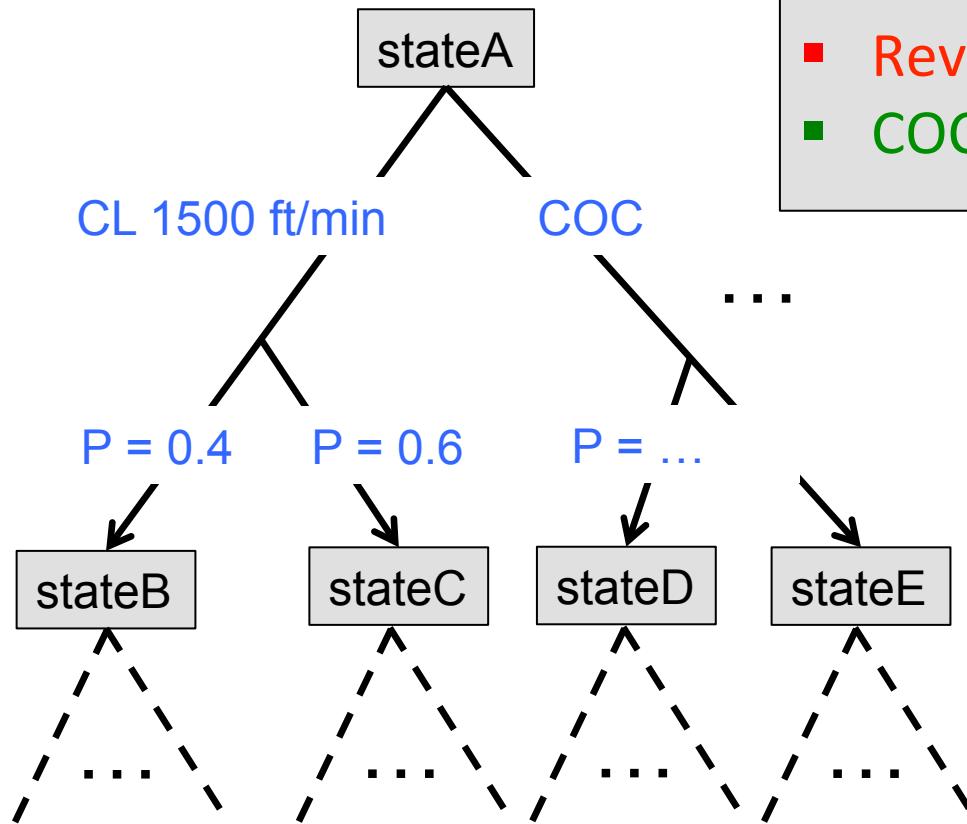
- 40 secs from Near-Mid-Air Collision (NMAC)
- state variables
 - h_{rel} : relative altitude, in [-1000...1000] *ft*
 - $dh_{\text{own}} / dh_{\text{int}}$: ownship / intruder climb rates, in [-2500...2500] *ft/min*
- $a_{\text{prev}} / s_{\text{RA}}$: advisory issued by ACAS X in previous sec / current pilot state, both in {COC, CL/DES1500, SCL/SDES1500, SCL/SDES2500}
- update and advisory **frequency** is set to 1 sec
- **discretization resolution** n for a variable V means that V is discretized to n points above and n points below 0 within its interval of values. For example, discretization resolution of 10 for h_{rel} means:
 $\{-1000, -900, -800, \dots, 0, 100, \dots, 900, 1000\}$



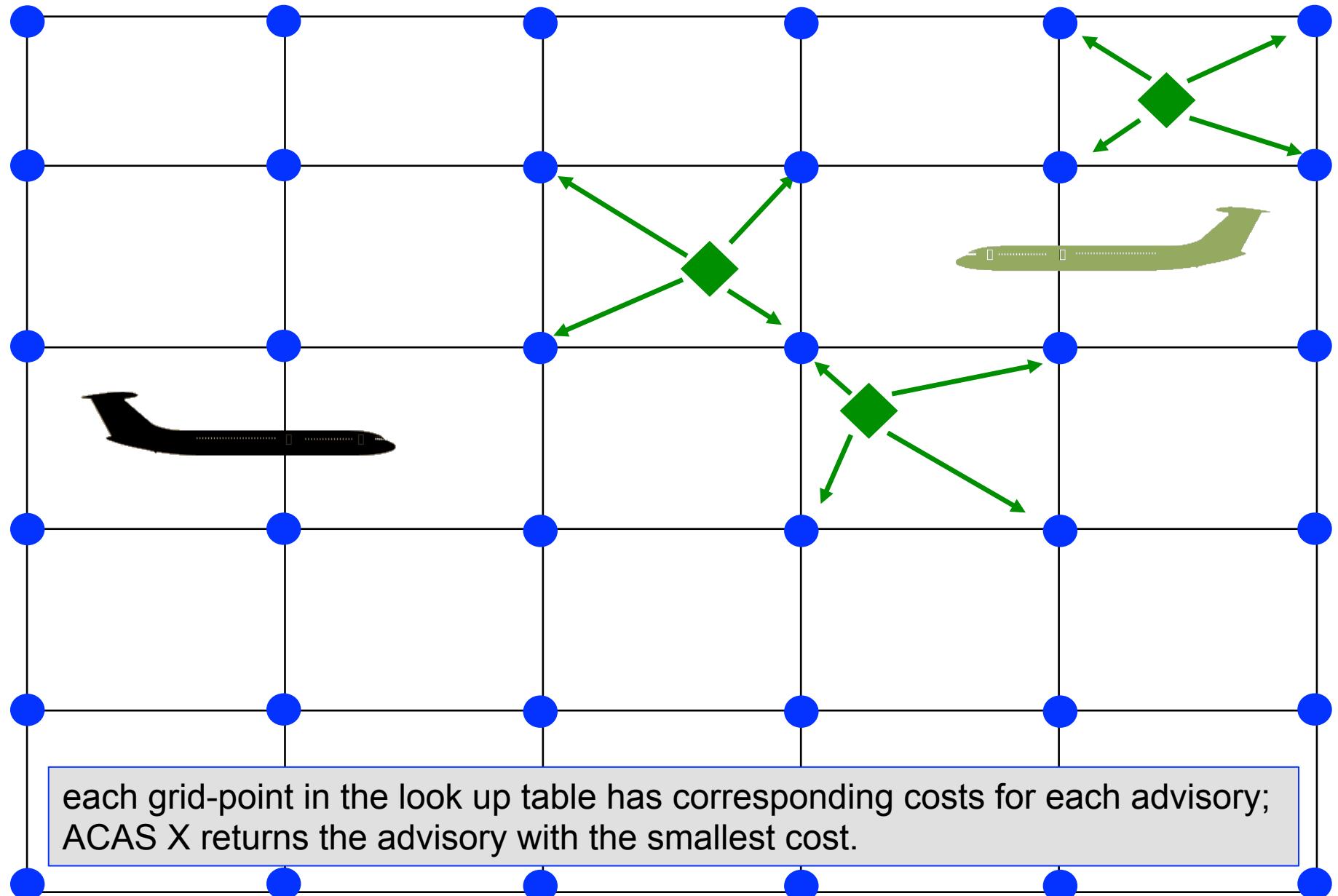
ACAS X – goals

minimize costs / maximize rewards

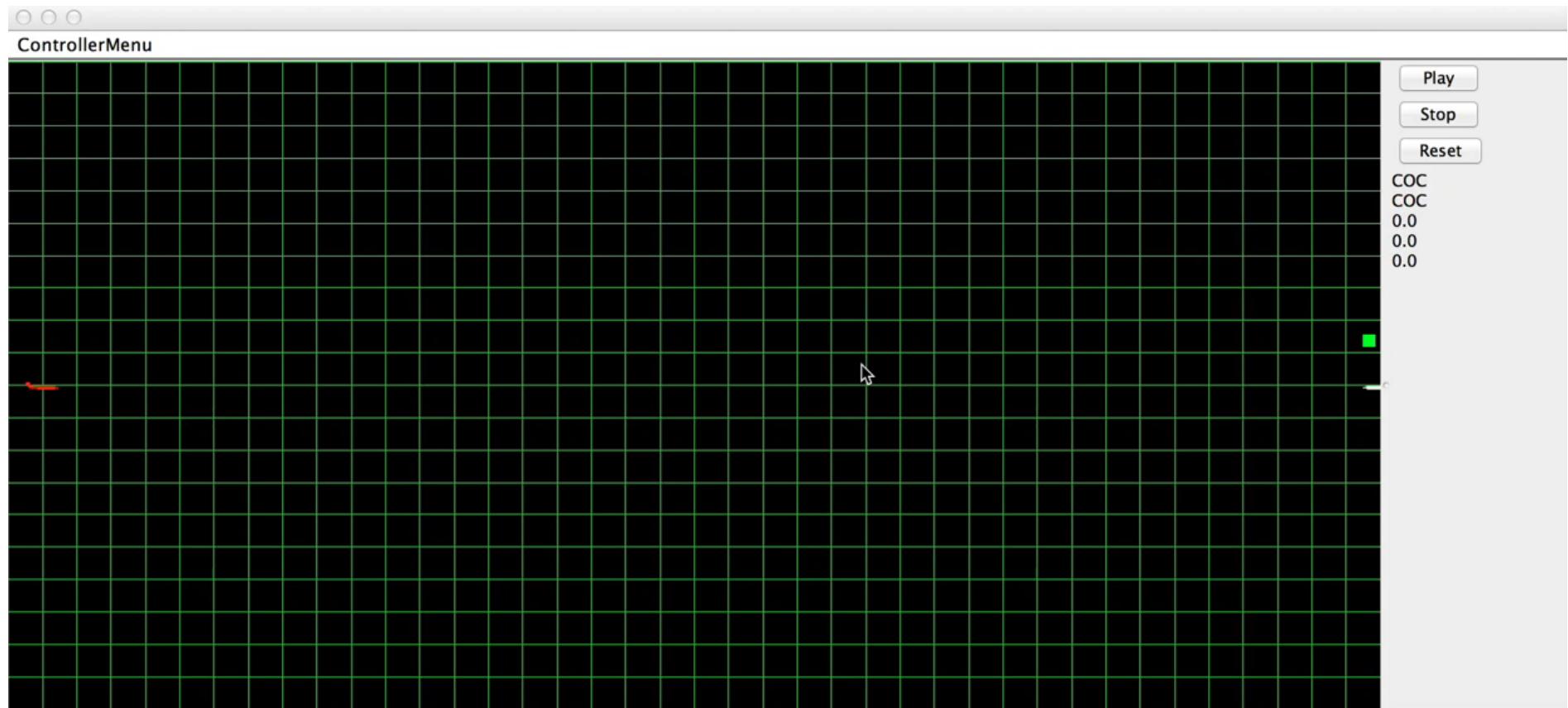
- NMAC (near-mid-air collision)
- Alert (from COC to advisory)
- Strengthening (strengthen advisory)
- Reversal (e.g. climb to descend)
- COC (clear of conflict)



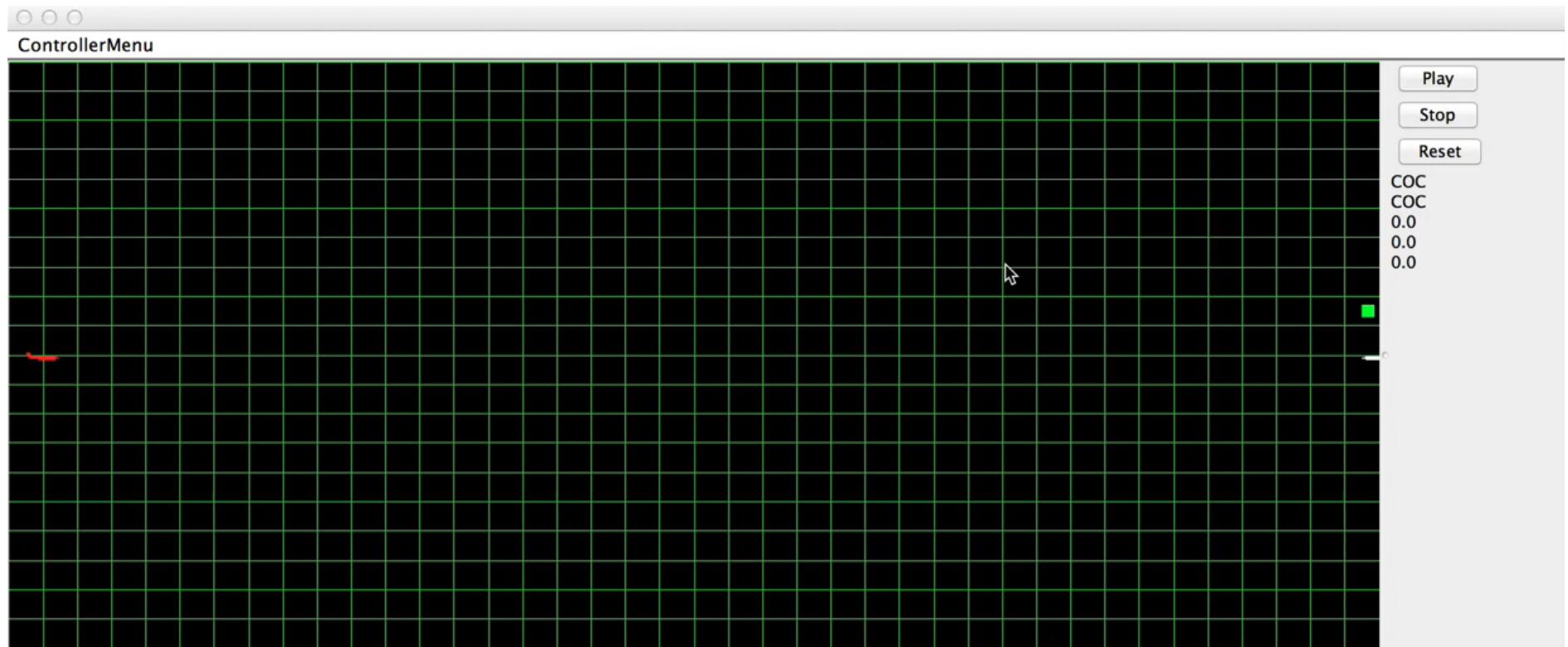
deploying ACAS X as a look up table



simulation with low NMAC weight (0.01)

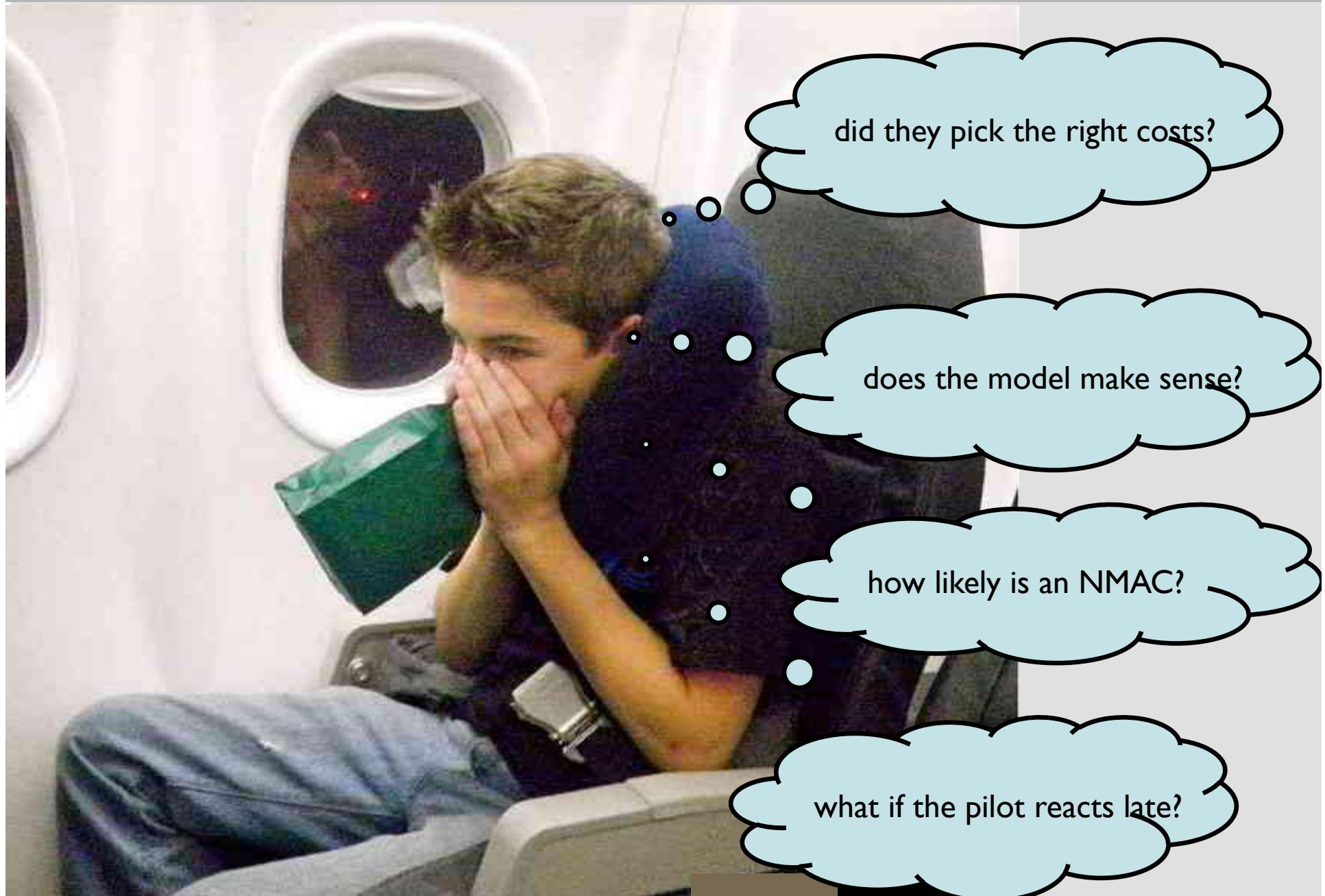


simulation with high NMAC weight (100)



verification starts with asking questions

would you trust ACAS X in a flight?

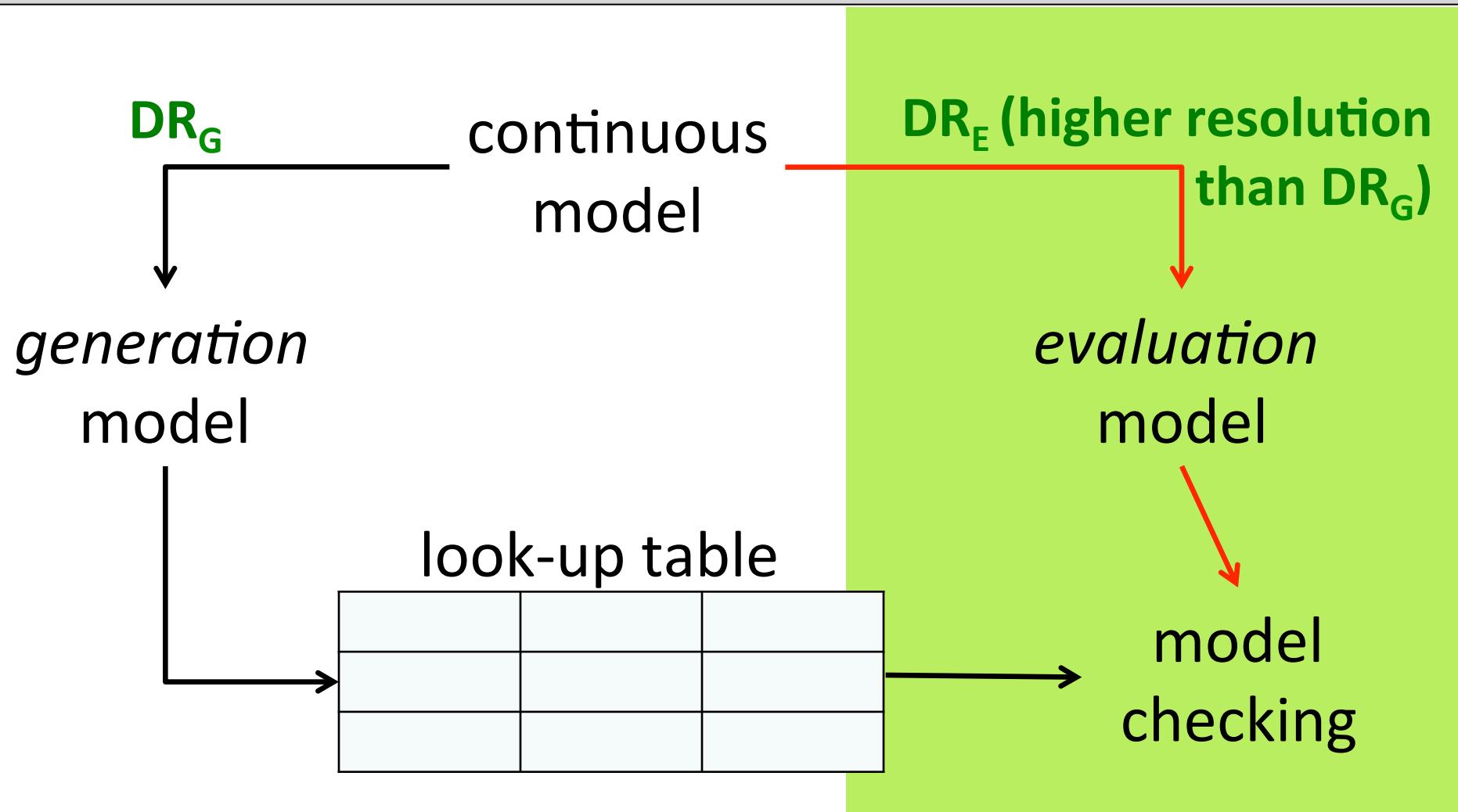


...and proceeds with answering them

discretization resolution (dh_{own} , dh_{int} , h_{rel})

DR_G : model discretization resolution for look-up table generation;
baseline [KC 2011] resolution is ($dh_{\text{own}}=10$, $dh_{\text{int}}=10$, $h_{\text{rel}}=10$)

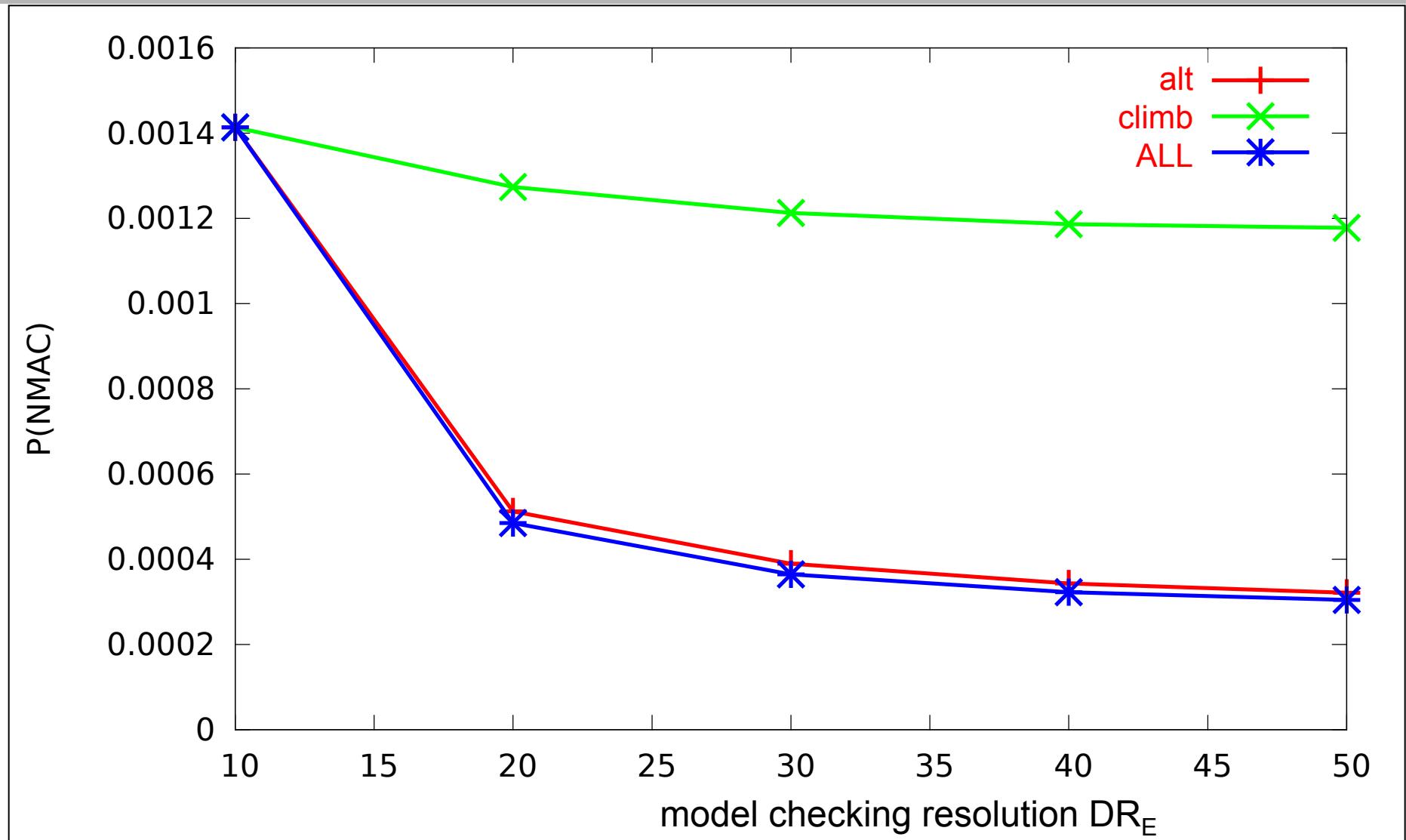
DR_E : model discretization resolution to model check look-up table



effects of resolution DR_E on model checking

- we compute $P(\text{NMAC})$ of the baseline look up table deployed in models that are obtained through discretization with varying resolutions DR_E (dh_{own} , dh_{int} , h_{rel})
 - **ALL** varies climb rates and relative altitude in DR_E : (20, 20, 20), (30, 30, 30), ...
 - **climb** varies climb rates only in DR_E : (20, 20, 10), (30, 30, 10), ...
 - **alt** varies relative altitude only in DR_E : (10, 10, 20), (10, 10, 30), ...

effects of resolution DR_E on model checking



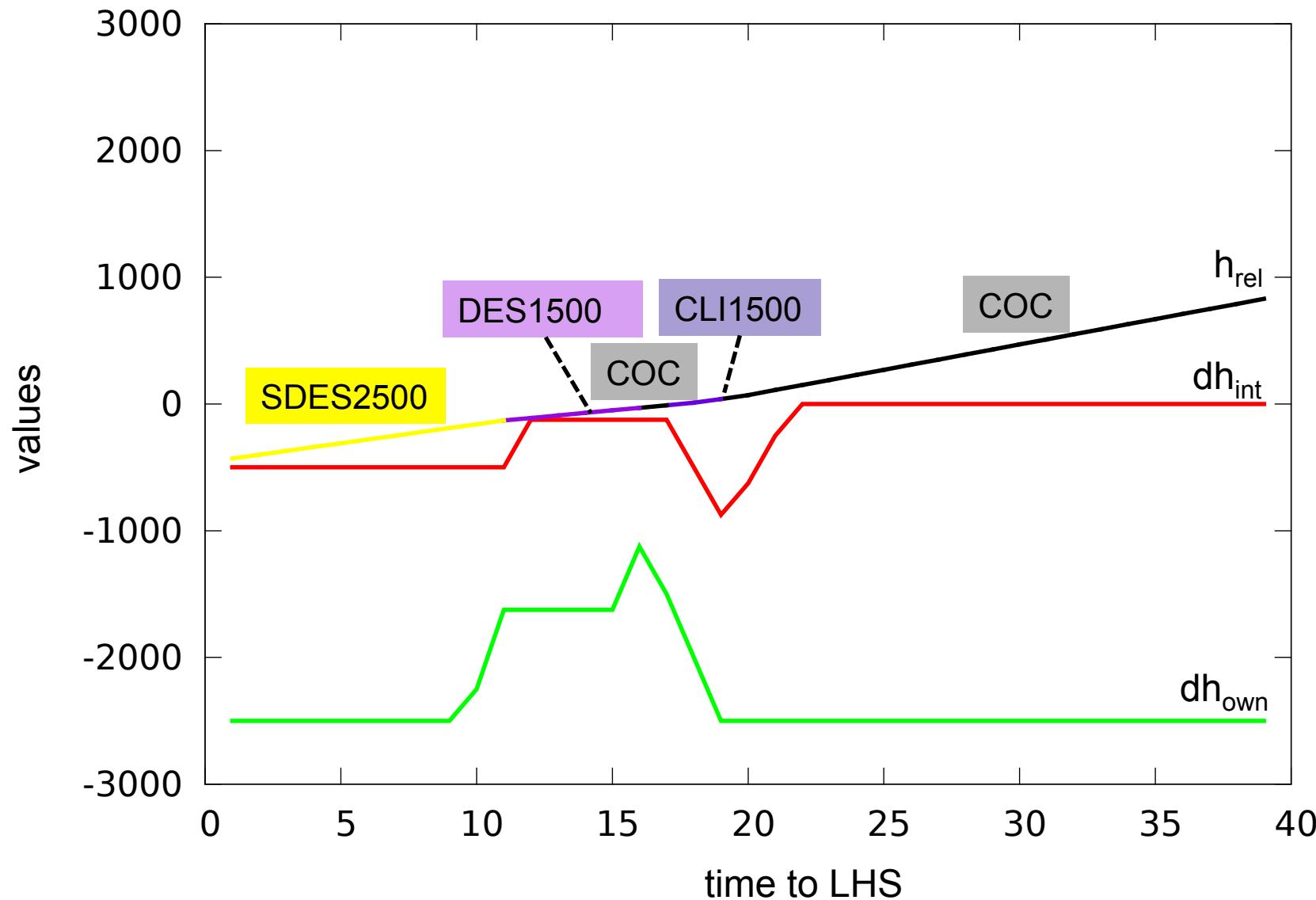
- $P(\text{NMAC})$ decreases with higher evaluation resolutions
- relative altitude discretization is indicative

probabilistic model checking $DR_E = (50, 50, 100)$

allows precise *automated analysis* of probabilistic properties expressed in a formal logic such as *PCTL*; generates *encounters* that exhibit property-related behaviors

- what is the probability of **NMAC**? $(P=?[\mathcal{F} \text{ NMAC}])$ 2.5×10^{-4}
- what if pilot responds immediately?
 $(P=?(\mathcal{F} \text{ NMAC} \mid \mathbf{G} a_{\text{prev}} = s_{\text{RA}}))$ 2.3×10^{-8}
- what is the probability of a **split advisory**? 1.8×10^{-3}
 $P=?[\mathcal{F} (\mathbf{!} \text{COC} \wedge \mathbf{P=1}[\mathbf{X} \text{COC}] \wedge \mathbf{P>0} [\mathcal{F} \mathbf{!} \text{COC}])]$
- split advisories are harder to directly take into account during look up table generation because they require to record history

split advisory encounter

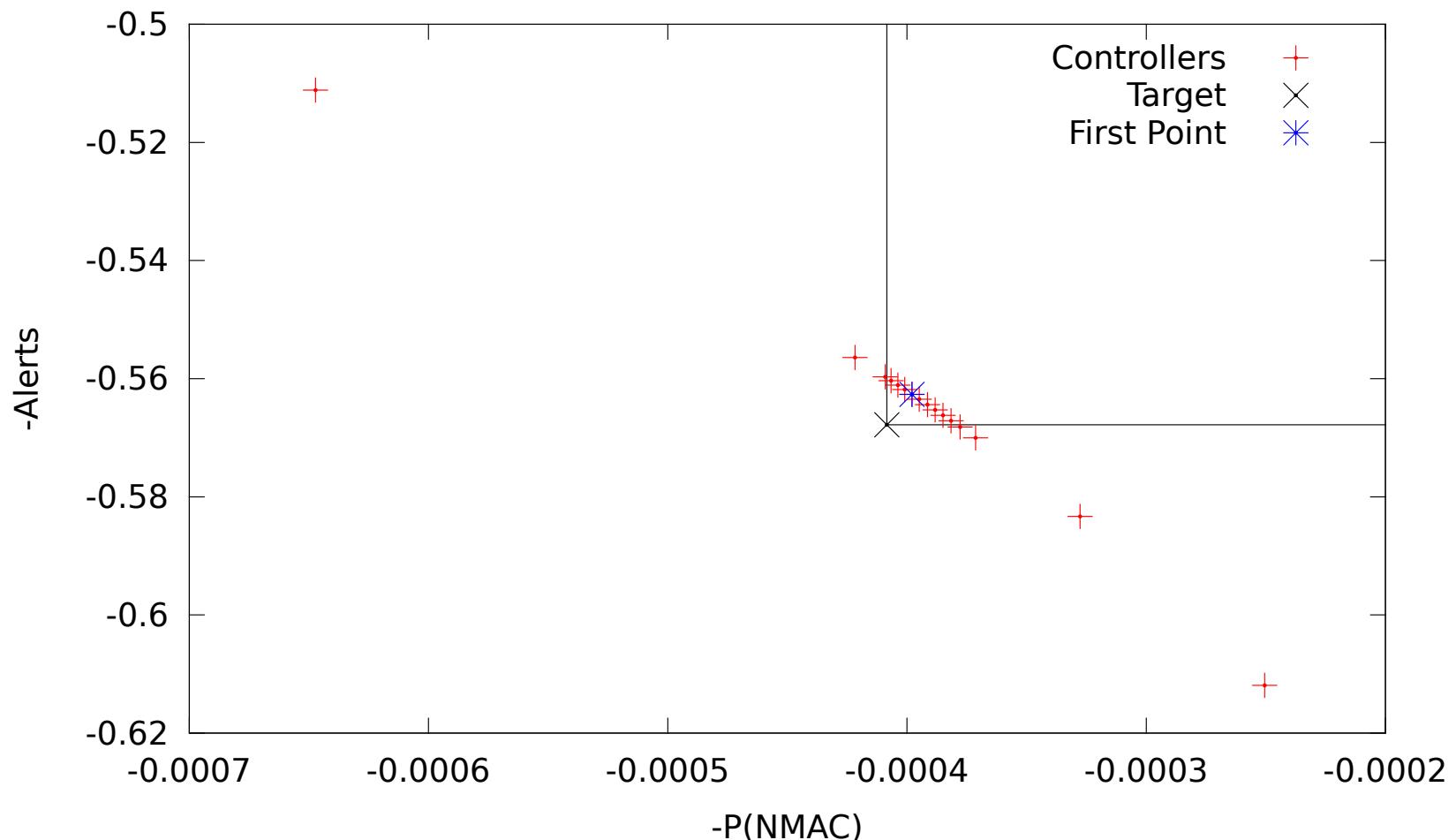


(reward for COC + cost of alert) < cost of reversal ("sneaky" reversals)

synthesis / design

weights vs. performance

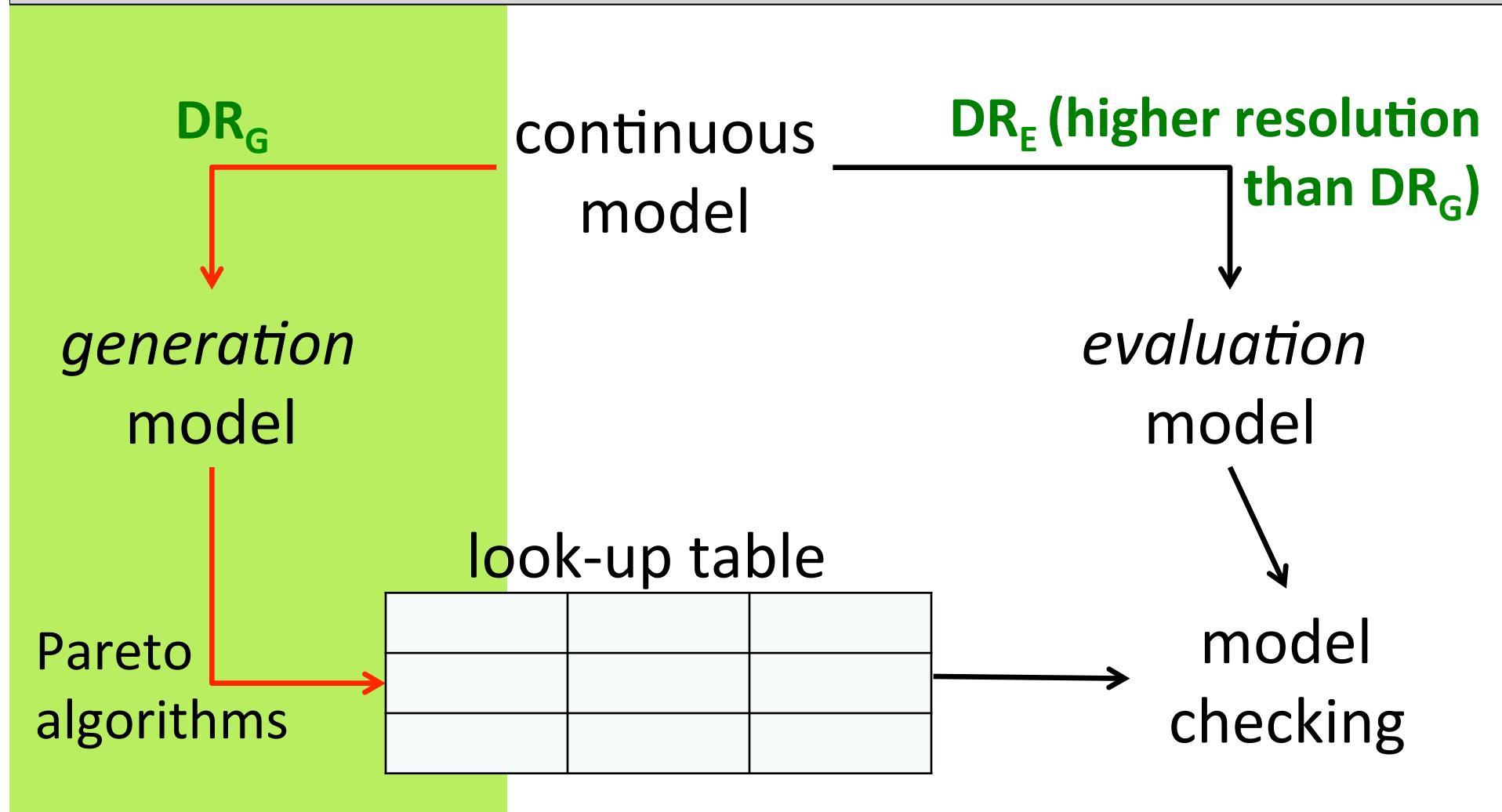
- tune look-up tables based on minimum acceptable performance
 - deterministic look-up tables based on weights form a convex Pareto front; we implement algorithms that approximate it *above* target performance



discretization resolution (dh_{own} , dh_{int} , h_{rel})

DR_G : model discretization resolution for look-up table generation;
baseline [KC 2011] resolution is ($dh_{own}=10$, $dh_{int}=10$, $h_{rel}=10$)

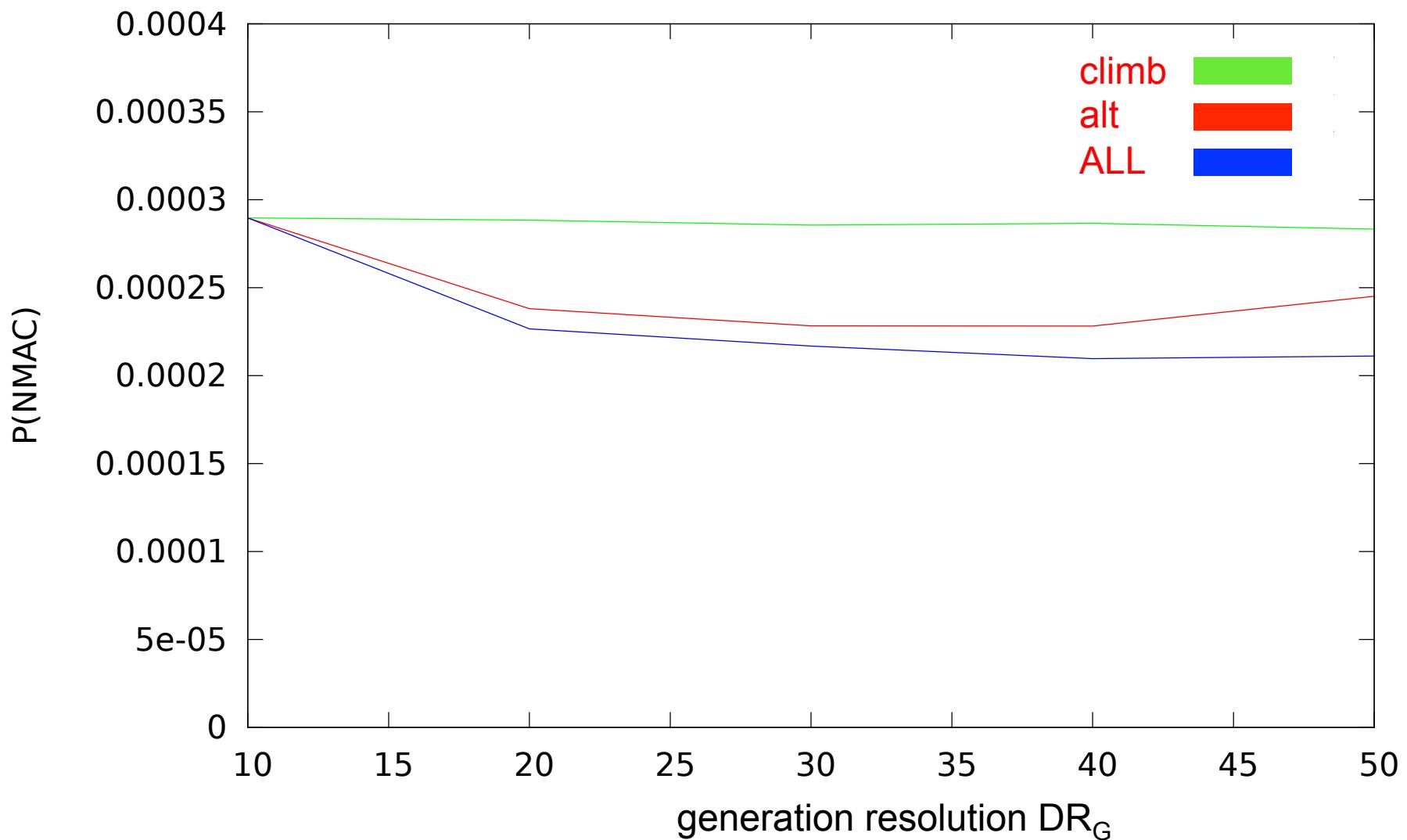
DR_E : model discretization resolution to model check look-up table



synthesizing better tables in higher DR_G

- question: what is the effect of resolution discretization DR_G on look-up table synthesis?
- experiment set up:
 - evaluate baseline ($dh_{own}=10, dh_{int}=10, h_{rel}=10$) [KC 2011] in new DR_G
 - use result as target for synthesis
- how we vary resolutions DR_G ($dh_{own}, dh_{int}, h_{rel}$)
 - **ALL** varies climb rates and relative altitude in DR_G : (20, 20, 20), (30, 30, 30), ...
 - **climb** varies climb rates only in DR_G : (20, 20, 10), (30, 30, 10), ...
 - **alt** varies relative altitude only in DR_G : (10, 10, 20), (10, 10, 30), ...
 - *note that **alt** results in the smallest look up tables – in terms of numbers of states – for each value increase*
- compare the synthesized look-up tables in $DR_E = (50, 50, 100)$

table synthesis at different resolutions



recommendation: (10, 10, 30), or (n, n, 3*n)

verification achievements

- we could not use off-the-shelf tools, so we built **VeriCA** toolset
 - our tools support models written in Java
 - we customized verification and synthesis algorithms for ACAS X needs
- we analyzed ACAS X version that we reproduced based on:
Kochenderfer, M. J., and Chryssanthacopoulos, J. P. Robust airborne collision avoidance through dynamic programming. Project Report ATC-371, Massachusetts Institute of Technology, Lincoln Laboratory, 2011.
- **ETAPS 2014 EASST best paper award**
 - Christian von Essen, Dimitra Giannakopoulou: *Analyzing the Next Generation Airborne Collision Avoidance System*, TACAS 2014.
- **FAA / NASA Ames Interagency Agreement** for ACAS X and VeriCA
 - thus able to apply our subsequent work on the actual ACAS X code

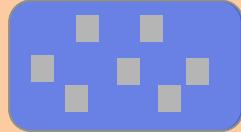
model quality

comparing model to real world

State machine model with probabilistic transitions (MDP) is used to generate onboard look-up table. *The MDP is not present in the onboard system.*

We **defined** Conformance Relations that compare MDP model to flight data

t



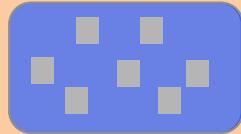
state estimate at time t
on look-up table

comparing model to real world

State machine model with probabilistic transitions (MDP) is used to generate onboard look-up table. *The MDP is not present in the onboard system.*

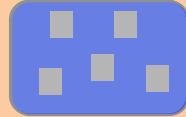
We **defined** Conformance Relations that compare MDP model to flight data

t



state estimate at time t
on look-up table

t+1sec



state estimate at time t+1
on look-up table

comparing model to real world

State machine model with probabilistic transitions (MDP) is used to generate onboard look-up table. *The MDP is not present in the onboard system.*

We **defined** Conformance Relations that compare MDP model to flight data

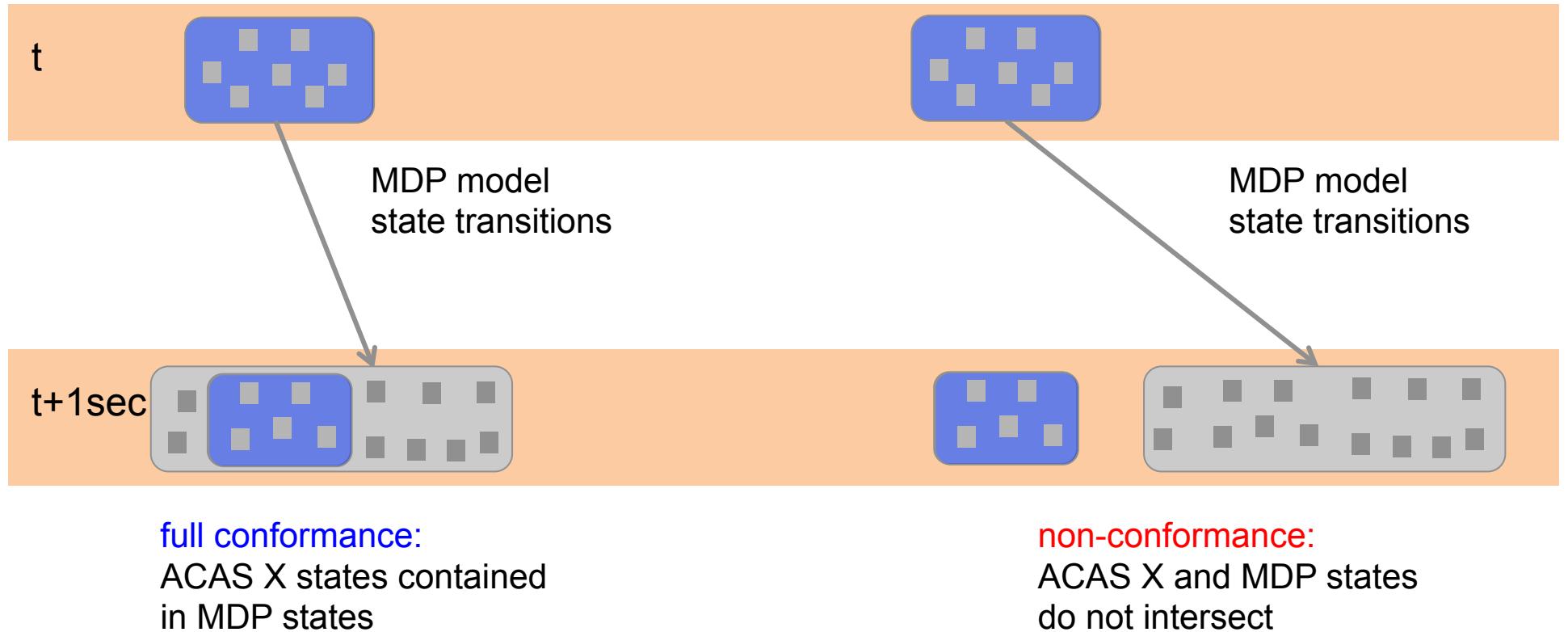


full conformance:
ACAS X states contained
in MDP states

comparing model to real world

State machine model with probabilistic transitions (MDP) is used to generate onboard look-up table. *The MDP is not present in the onboard system.*

We **defined** Conformance Relations that compare MDP model to flight data

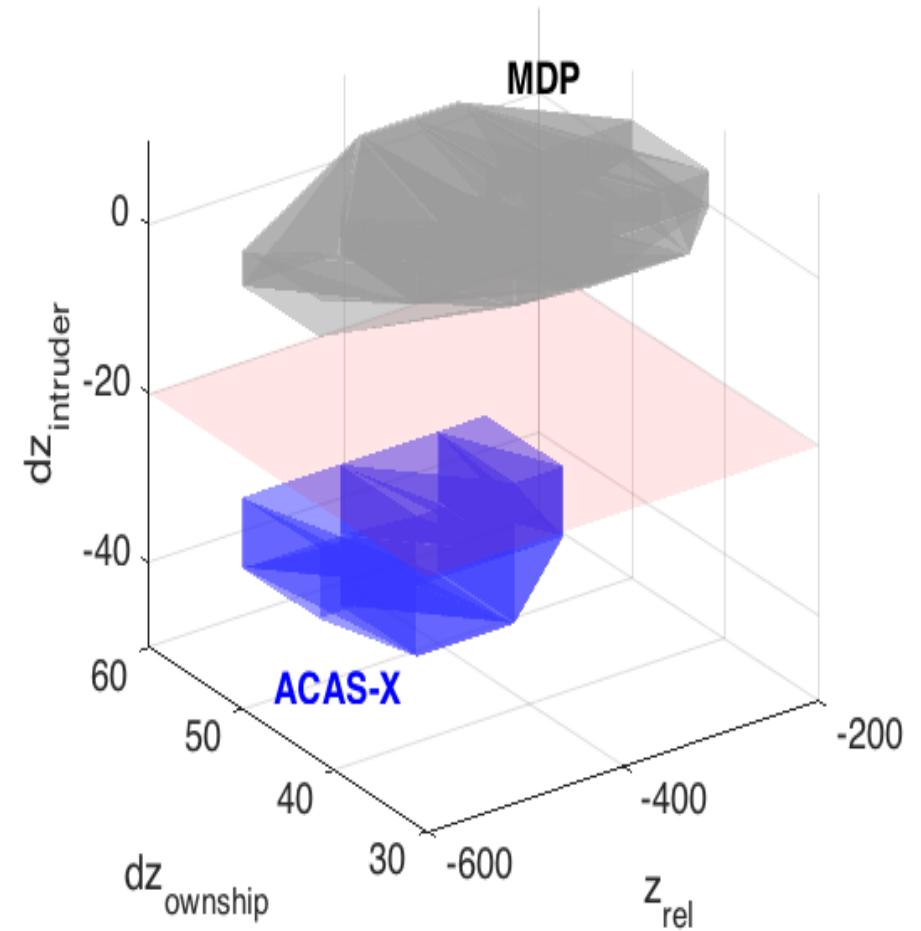


data generation and analysis

Data generation: Non-conforming encounters are rare in test data of the ACAS X distribution. We used a reinforcement learning framework to target generation of non-conforming simulated encounters.

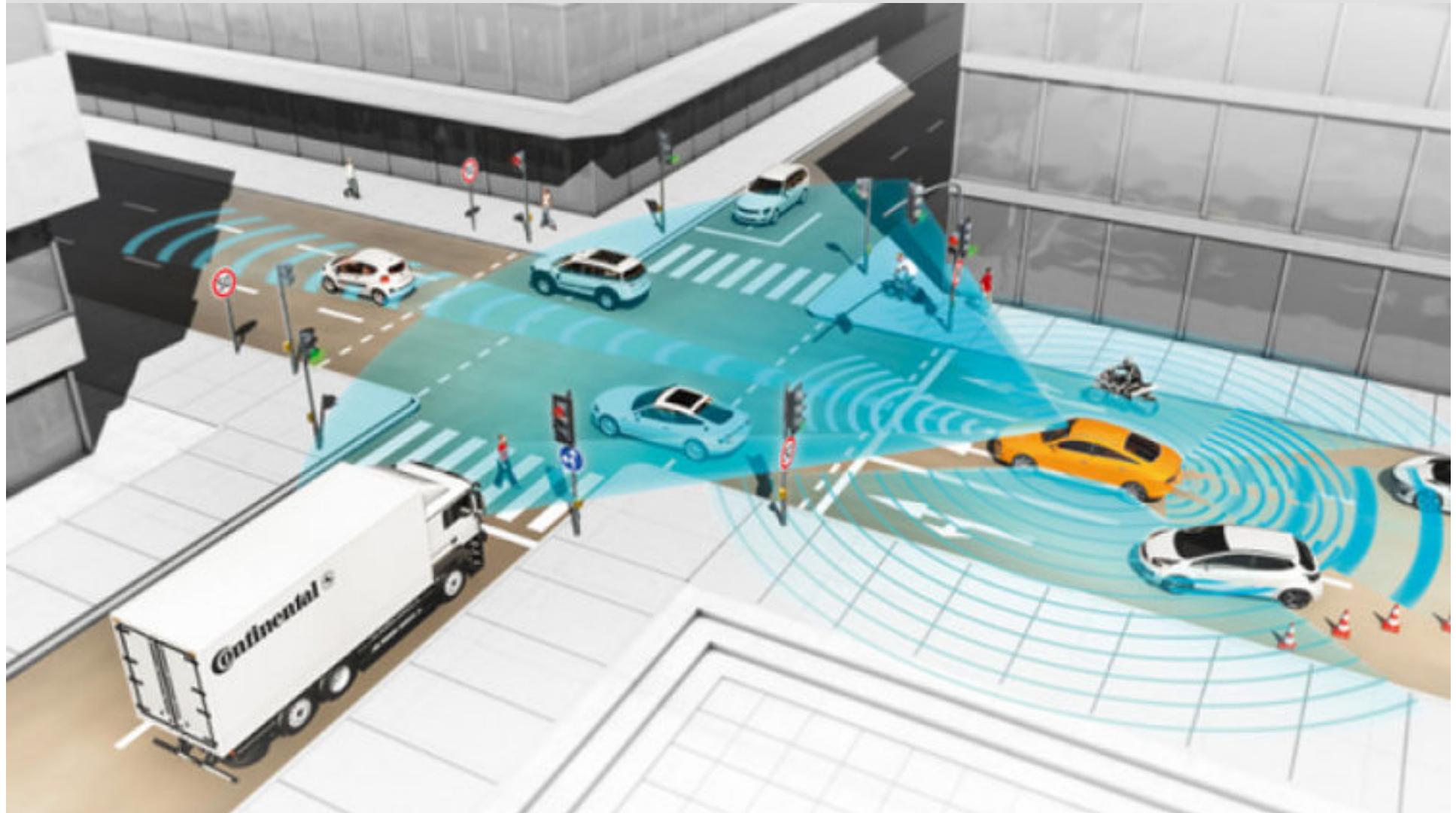
Data Analysis: The intruder climb rate has been identified as a common factor for divergence across the data. *Further analysis is needed.*

Open question: Does non-conformance imply potential violation of safety requirements?



applicability

self-driving cars

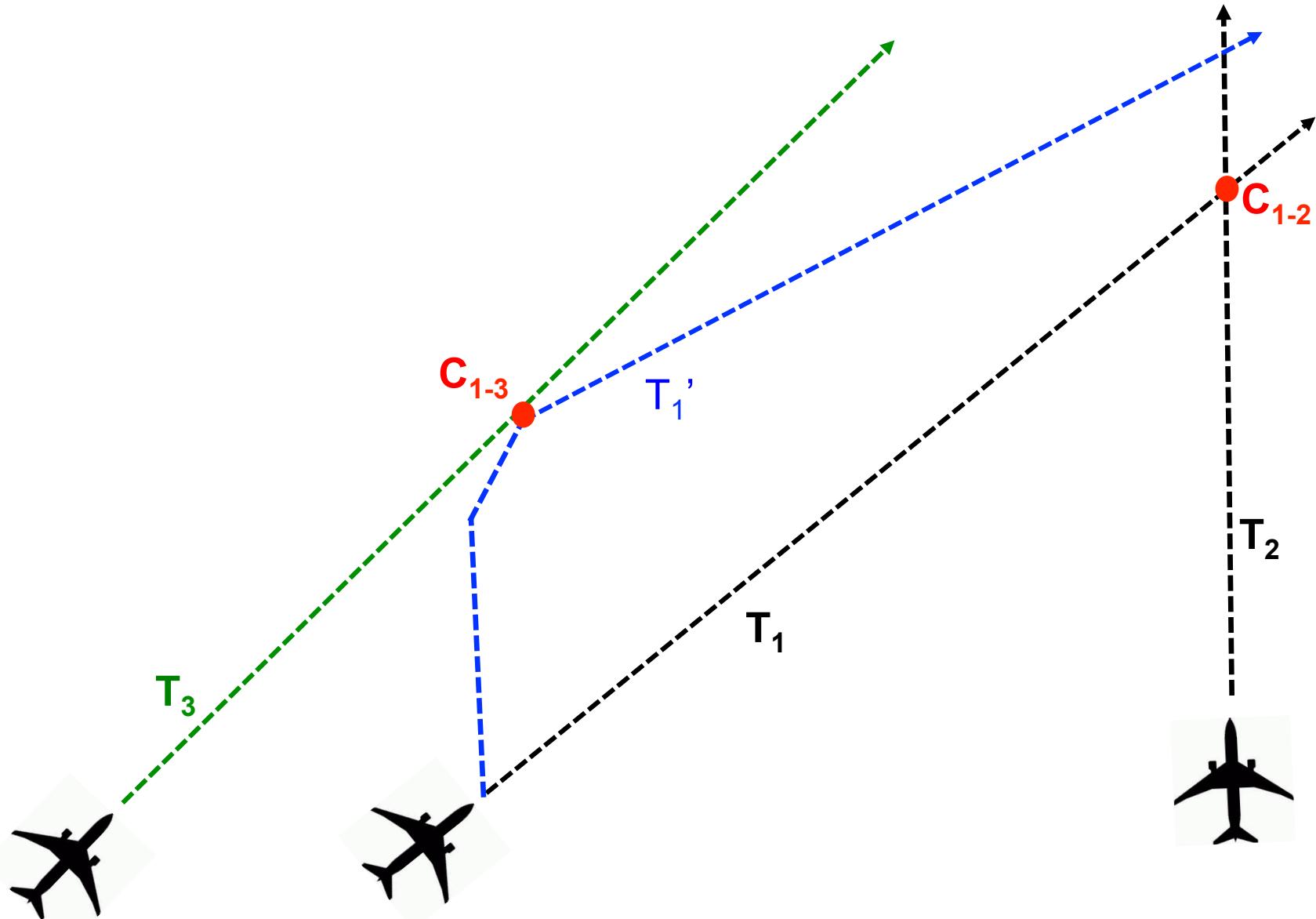


V&V of autonomy

- formulation of requirements is harder – autonomy-specific?
 - optimization, adaptive and learning algorithms
 - example: loss of separation, ACAS X

generate strategic secondary aircraft

no picked resolution is allowed to cause a more imminent secondary conflict



V&V of autonomy

- formulation of requirements is harder – autonomy-specific?
 - optimization, adaptive and learning algorithms
 - example: separation assurance, ACAS X
- need for advanced testing infrastructures
 - test case generation for stress-testing and requirements coverage
 - examples: ACAS X, separation assurance, autonomous vehicles
- V&V at runtime, including requirements
 - ACAS X (error prediction with statistical learning)
 - separation assurance
- trust
 - extensive verification
 - explanation of decision-making algorithms

Collaborators / Publications

1. Dimitra Giannakopoulou, David H. Bushnell, Johann Schumann, Heinz Erzberger, Karen Heere: **Formal testing for separation assurance.** Ann. Math. Artif. Intell. 63(1): 5-30 (2011)
2. Dimitra Giannakopoulou, Falk Howar, Malte Isberner, Todd Lauderdale, Zvonimir Rakamaric, Vishwanath Raman: **Taming test inputs for separation assurance.** ASE 2014.
3. Marko Dimjasevic, Dimitra Giannakopoulou: **Test-Case Generation for Runtime Analysis and Vice Versa: Verification of Aircraft Separation Assurance.** ISSTA2015.
4. Christian von Essen, Dimitra Giannakopoulou: **Probabilistic verification and synthesis of the next generation airborne collision avoidance system.** STTT 18(2): 227-243 (2016). Extended version of TACAS 2014 paper awarded ETAPS 2014 EASST best paper.
5. Dimitra Giannakopoulou, Dennis Guck, Johann Schumann: **Exploring Model Quality for ACAS X.** FM 2016.